Correlation functions for symmetrized increasing subsequences

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Abstract

We show that the correlation functions associated to symmetrized increasing subsequence problems can be expressed as pfaffians of certain antisymmetric matrix kernels, thus generalizing the result of [11] for the unsymmetrized case.

Introduction

In [11], Okounkov derived the following symmetric function identity: For any finite subset $S \subset \mathbb{Z}$,

$$\sum_{\lambda:S\subset \{\lambda_j-j:j\in\mathbb{Z}^+\}} s_\lambda(x)s_\lambda(y) = \sum_\lambda s_\lambda(x)s_\lambda(y)\det(K(S)),$$

where K(S) is the appropriate principal minor of an explicit infinite matrix K, and λ ranges over partitions. The main applications of this result are to the asymptotic analysis of generalized increasing subsequence problems; such a problem induces a distribution on partitions such that λ occurs with probability $s_{\lambda}(x)s_{\lambda}(y)$, appropriately specialized (see Section 7 of [3]). For instance, the distribution of the kth row of λ can be computed from this result in terms of a certain Fredholm determinant.

In [3], [4], [5], we considered five classes of generalized increasing subsequence problems, corresponding to different choices of symmetry imposed on the problem. As the above result only applies to the symmetry-free class \square , it is natural to wonder whether analogous results hold in the other cases. As we shall see in the present note, there is a matrix associated to each of the five symmetry classes such that the corresponding correlation functions are given as either the determinant or the pfaffian of appropriate minors. Each of these symmetry classes corresponds to an appropriate Cauchy-Littlewood type identity; using the present techniques, we can obtain analogous results for the remaining three Littlewood identities (see Section 7).

We begin in Section 1 by giving a fairly general theorem (Theorem 1.1), inspired by the results of [13], to the effect that for any measure space (X, λ) and any probability distribution on X^{2m} with density of the form

$$\det(\phi_j(x_k)) \operatorname{pf}(\epsilon(x_j, x_k)),$$

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the corresponding correlation function can be expressed as a pfaffian. Since the distributions we are interested in are not of this form, we cannot directly apply Theorem 1.1. However, in each case, we can write the desired correlation function as a formal limit of correlation functions to which Theorem 1.1 does apply. Section 2 gives some lemmas on formal inverses of infinite matrices which we use in sections 3 through 7 to simplify the obtained pfaffian kernels. Finally, in section 8, we discuss the analogue for pfaffians of the notion of Fredholm determinant, and give a Fredholm pfaffian-based derivation of Theorem 1.1.

For the (somewhat involved) definitions of the increasing subsequence problems considered below, we refer the reader to Section 7 of [3]; we will also use the somewhat more general notion of parameter set introduced in [12].

1 Correlation functions as pfaffians

The correlation functions we will be studying below can all be expressed as pfaffians of certain antisymmetric matrix kernels. Recall that a matrix kernel on a space X is a matrix-valued function on $X \times X$; for a matrix kernel K, we define its transpose K^t by

$$K^{t}(x,y) = K(y,x)^{t}. \tag{1.1}$$

Given a finite sequence $\Sigma = x_1, x_2, \dots x_k$ of elements of X, the restriction $K(\Sigma)$ of K to Σ is defined to be the block matrix with ijth block $K(x_i, x_j)$; note that $K^t(\Sigma) = K(\Sigma)^t$. In particular, if K is antisymmetric, then so is $K(\Sigma)$, and thus we can compute the pfaffian pf $(K(\Sigma))$. When K is even-dimensional, this is invariant under reordering of Σ , and thus depends only on the underlying set. For a finite subset $S \subset X$, we define pf(K(S)) accordingly. By convention, the pfaffian of a 0×0 matrix is 1, so pf $(K(\emptyset)) = 1$. Given two sequences Σ_{\pm} , we define $K(\Sigma_+, \Sigma_-)$ in the obvious way, and write $K(S_+, S_-)$ for sets S_{\pm} whenever the meaning is clear. Thus, for instance, if S_+ and S_- are disjoint, we can write

$$pf(K(S_{+} \cup S_{-})) = pf\begin{pmatrix} K(S_{+}, S_{+}) & K(S_{+}, S_{-}) \\ K(S_{-}, S_{+}) & K(S_{-}, S_{-}) \end{pmatrix}.$$
(1.2)

We also adopt corresponding notations for determinants.

The way in which such pfaffians arise in the sequel is via the following theorem:

Theorem 1.1. Let (X, λ) be a measure space, let $\phi_1, \ldots, \phi_{2m}$, be functions from X to \mathbb{C} , let ϵ be an antisymmetric function from $X \times X$ to \mathbb{C} , and assume the antisymmetric matrix

$$M_{jk} = \int_{x,y \in X} \phi_j(x)\epsilon(x,y)\phi_k(y)\lambda(dx)\lambda(dy)$$
(1.3)

is well-defined and invertible. For a finite subset $S = \{x_1, x_2, \dots x_l\} \subset X$ with $l \leq 2m$, we define a correlation function

$$R(S; \phi, \epsilon) := \frac{1}{(2m-l)! \operatorname{pf}(M)} \int_{x_{l+1}, \dots x_{2m} \in X} \det(\phi_j(x_k)) \operatorname{pf}(\epsilon(x_j, x_k)) \prod_{l+1 \le j \le 2m} \lambda(dx_j); \tag{1.4}$$

for |S| > 2m, we set $R(S; \phi, \epsilon) = 0$. Then $R(S; \phi, \epsilon) = \operatorname{pf}(K(S))$, where K is the antisymmetric matrix kernel

$$K(x,y) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \phi_k(y) & \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} (\epsilon \cdot \phi_k)(y) \\ \sum_{1 \le j,k \le 2m} (\epsilon \cdot \phi_j)(x) M_{jk}^{-t} \phi_k(y) & -\epsilon(x,y) + \sum_{1 \le j,k \le 2m} (\epsilon \cdot \phi_j)(x) M_{jk}^{-t} (\epsilon \cdot \phi_k)(y) \end{pmatrix}, \tag{1.5}$$

and for a function $f: X \to \mathbb{C}$,

$$(\epsilon \cdot f)(x) = \int_{y \in X} \epsilon(x, y) f(y) \lambda(dy). \tag{1.6}$$

Proof. We first consider the case $|S| \ge 2m$. In that case, if the matrix $\Phi := \phi_j(S)$ is singular, then the odd rows of K(S) are linearly dependent and thus pf(K(S)) = 0. We may thus assume |S| = 2m and Φ is nonsingular. Then we can express $(\epsilon \cdot \phi_j)(x)$ on S as a linear combination of the functions $\phi_j(x)$. Using this we find that

$$pf(K(S)) = pf(K'(S)), \tag{1.7}$$

where

$$K'(x,y) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \phi_k(y) & 0\\ 0 & -\epsilon(x,y) \end{pmatrix}.$$
 (1.8)

But then

$$\operatorname{pf}(K'(S)) = \operatorname{pf}(\Phi M^{-t}\Phi^{t})\operatorname{pf}(\epsilon(x_{i}, x_{k})) = \operatorname{pf}(M)^{-1}\operatorname{det}(\phi_{i}(x_{k}))\operatorname{pf}(\epsilon(x_{i}, x_{k})), \tag{1.9}$$

as required.

Now, suppose we know the theorem for sets of size $\geq l$, and let S be a set of size l-1. Then

$$R(S; \phi, \epsilon) = \frac{1}{2m - l + 1} \int_{x_l \in X} R(S \cup \{x_l\}; \phi, \epsilon) \lambda(dx_l) = \frac{1}{2m - l + 1} \int_{x_l \in X} \operatorname{pf}(K(S \cup \{x_l\})) \lambda(dx_l)$$
(1.10)

It thus suffices to show

$$\int_{x_l \in X} \operatorname{pf}(K(S \cup \{x_l\})) \lambda(dx_l) = (2m - l + 1) \operatorname{pf}(K(S)).$$
(1.11)

Expand pf $(K(S \cup \{x_l\}))$ along the bottom two rows and integrate, then simplify using the following integrals:

$$\int_{x_l \in X} K(x_l, x_l)_{21} \lambda(dx_l) = -2m \tag{1.12}$$

$$\int_{x_l \in X} K(x_l, x_j)_{11} K(x_l, x_k)_{21} \lambda(dx_l) = K(x_j, x_k)_{11}$$
(1.13)

$$\int_{x_l \in X} K(x_l, x_j)_{12} K(x_l, x_k)_{21} \lambda(dx_l) = K(x_j, x_k)_{21}$$
(1.14)

$$\int_{x_l \in X} K(x_l, x_j)_{11} K(x_l, x_k)_{22} \lambda(dx_l) = 0$$
(1.15)

$$\int_{x_l \in X} K(x_l, x_j)_{12} K(x_l, x_k)_{22} \lambda(dx_l) = 0$$
(1.16)

We thus see that the 22 terms contribute nothing. For the 21 terms, $K(x_l, x_l)_{21}$ contributes $2m \operatorname{pf}(K(S))$ directly, while the terms associated to $K(x_l, x_k)_{21}$ give precisely the expansion of $\operatorname{pf}(K(S))$ along the first x_k column, up to an overall sign change. We thus obtain a total of $2m \operatorname{pf}(K(S)) - (l-1) \operatorname{pf}(K(S))$, as required. \square

Remark 1. The above operator essentially appeared in [13], which considered the case $\phi_j \propto x^{j-1}$, $\epsilon(x,y) =$ $\frac{1}{2}$ sgn(y-x); that reference did not obtain a direct formula for the correlation functions, however. See Section 8 for a derivation of the theorem along their lines. The above proof generalizes that used (for the same special case) in [10], Chapter 6. Note that in [10], the correlation functions are stated as "quaternion determinants", essentially the restriction of the notion of pfaffian to block matrices.

Remark 2. When $S = \emptyset$, we find

$$\frac{1}{(2m)!} \int_{x_1, \dots x_{2m} \in X} \det(\phi_j(x_k)) \operatorname{pf}(\epsilon(x_j, x_k)) \prod_{1 \le j \le 2m} \lambda(dx_j) = \operatorname{pf}(M), \tag{1.17}$$

proving a result of [8].

Remark 3. The kernel K is, of course, not unique; for instance, we may use $K'(x,y) = T(x)K(x,y)T(y)^t$ where T is any function from X to $SL_2(\mathbb{C})$.

Corollary 1.2. Let (X,λ) and (Y,μ) be measure spaces, let ϕ_1,\ldots,ϕ_{2m} be measurable functions from $X\to\mathbb{C}$, let $\psi_1, \ldots \psi_{2m}$ be measurable functions from $Y \to \mathbb{C}$, and let κ be a function from $X \times Y$ to \mathbb{C} . Assume that the antisymmetric matrix

$$M_{jk} = \int_{x \in X, y \in Y} (\phi_j(x)\psi_k(y) - \phi_k(x)\psi_j(y))\kappa(x, y)\lambda(dx)\mu(dy)$$
(1.18)

is well-defined and invertible. Then, for finite sets $S_0 = \{x_1, x_2, \dots x_{l_0}\} \subset X$, $S_1 = \{y_1, y_2, \dots y_{l_1}\} \subset Y$, defined $S_1 = \{y_1, y_2, \dots y_{l_1}\} \subset Y$, defined $S_2 = \{y_1, y_2, \dots y_{l_1}\} \subset Y$, defined $S_3 = \{y_1, y_2, \dots y$

$$R(S_0, S_1; \phi, \psi, \kappa) = \frac{1}{(m - l_0)!(m - l_1)! \operatorname{pf}(M)} \int_{\substack{x_{l_0 + 1}, \dots x_m \in X \\ y_{l_1 + 1}, \dots y_m \in Y}} \det(\phi_j(x_k) \ \psi_j(y_k)) \det(\kappa(x_j, y_k))$$

$$\prod_{l_0 + 1 \le j \le m} \lambda(dx_j) \prod_{l_1 + 1 \le j \le m} \mu(dy_j), \quad (1.19)$$

we have

$$R(S_0, S_1; \phi, \psi, \kappa) = \operatorname{pf} \begin{pmatrix} K_{00}(S_0, S_0) & K_{01}(S_0, S_1) \\ K_{10}(S_1, S_0) & K_{11}(S_1, S_1) \end{pmatrix},$$
(1.20)

where

$$K_{00}(x,x') = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \phi_k(x') & \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} (\kappa \cdot \psi_k)(x') \\ \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_j)(x) M_{jk}^{-t} \phi_k(x') & \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_j)(x) M_{jk}^{-t} (\kappa \cdot \psi_k)(x') \end{pmatrix}$$
(1.21)

$$K_{01}(x,y) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t}(\kappa^t \cdot \phi_k)(y) & \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \psi_k(y) \\ \kappa(x,y) + \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_j)(x) M_{jk}^{-t}(\kappa^t \cdot \phi_k)(y) & \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_j)(x) M_{jk}^{-t} \psi_k(y) \end{pmatrix}$$
(1.22)

$$K_{01}(x,y) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_{j}(x) M_{jk}^{-t}(\kappa^{t} \cdot \phi_{k})(y) & \sum_{1 \le j,k \le 2m} \phi_{j}(x) M_{jk}^{-t}\psi_{k}(y) \\ \kappa(x,y) + \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_{j})(x) M_{jk}^{-t}(\kappa^{t} \cdot \phi_{k})(y) & \sum_{1 \le j,k \le 2m} (\kappa \cdot \psi_{j})(x) M_{jk}^{-t}\psi_{k}(y) \end{pmatrix}$$

$$K_{10}(y,x) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} (\kappa^{t} \cdot \phi_{j})(y) M_{jk}^{-t}\phi_{k}(x) & -\kappa(x,y) + \sum_{1 \le j,k \le 2m} (\kappa^{t} \cdot \phi_{j})(y) M_{jk}^{-t}(\kappa \cdot \psi_{k})(x) \\ \sum_{1 \le j,k \le 2m} \psi_{j}(y) M_{jk}^{-t}\phi_{k}(x) & \sum_{1 \le j,k \le 2m} \psi_{j}(y) M_{jk}^{-t}(\kappa \cdot \psi_{k})(x) \end{pmatrix}$$

$$K_{11}(y,y') = \begin{pmatrix} \sum_{1 \le j,k \le 2m} (\kappa^{t} \cdot \phi_{j})(y) M_{jk}^{-t}(\kappa^{t} \cdot \phi_{k})(y') & \sum_{1 \le j,k \le 2m} (\kappa^{t} \cdot \phi_{j})(y) M_{jk}^{-t}\psi_{k}(y') \\ \sum_{1 \le j,k \le 2m} \psi_{j}(y) M_{jk}^{-t}(\kappa^{t} \cdot \phi_{k})(y') & \sum_{1 \le j,k \le 2m} \psi_{j}(y) M_{jk}^{-t}\psi_{k}(y') \end{pmatrix}$$

$$(1.24)$$

$$K_{11}(y,y') = \begin{pmatrix} \sum_{1 \le j,k \le 2m} (\kappa^t \cdot \phi_j)(y) M_{jk}^{-t} (\kappa^t \cdot \phi_k)(y') & \sum_{1 \le j,k \le 2m} (\kappa^t \cdot \phi_j)(y) M_{jk}^{-t} \psi_k(y') \\ \sum_{1 \le j,k \le 2m} \psi_j(y) M_{jk}^{-t} (\kappa^t \cdot \phi_k)(y') & \sum_{1 \le j,k \le 2m} \psi_j(y) M_{jk}^{-t} \psi_k(y') \end{pmatrix}$$
(1.24)

for $x, x' \in X$, $y, y' \in Y$.

Proof. Define functions ϕ^+ on $X \uplus Y$ by

$$\phi_i^+(x) = \phi_j(x) \qquad \phi_i^+(y) = \psi_j(y)$$
 (1.25)

and an antisymmetric function ϵ on $(X \uplus Y)^2$ by

$$\epsilon(x, x') = 0$$
 $\epsilon(x, y) = \kappa(x, y)$ (1.26)

$$\epsilon(y, x) = -\kappa(x, y) \qquad \epsilon(y, y') = 0 \tag{1.27}$$

Then the function

$$\det(\phi_j^+(z_k))\operatorname{pf}(\epsilon(z_j, z_k)) \tag{1.28}$$

on $(X \uplus Y)^{2m}$ is 0 unless exactly half of the z_k are in Y, in which case it equals

$$\det(\phi_j(x_k) \ \psi_j(y_k)) \det(\kappa(x_j, y_k)). \tag{1.29}$$

Furthermore, the current matrix M is the same as the matrix associated to ϕ^+ and ϵ . We thus find that

$$R(S_0, S_1; \phi, \psi, \kappa) = R(S_0 \cup S_1; \phi^+, \epsilon),$$
 (1.30)

so we can apply Theorem 1.1; we compute

$$(\epsilon \cdot \phi^{+})(x) = (\kappa \cdot \psi)(x) \qquad (\epsilon \cdot \phi^{+})(y) = -(\kappa^{t} \cdot \phi)(y), \tag{1.31}$$

thus obtaining the desired result, up to transformation by

$$T(x) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \qquad T(y) = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}. \tag{1.32}$$

Remark. If Y = X, $\psi = \phi$, then we obtain a density on pairs of disjoint m-subsets of X. Taking the union, we obtain a density on 2m-subsets of X, which is of precisely the form considered in Theorem 1.1, with $\epsilon = \kappa - \kappa^t$. Thus the corollary may be viewed as a refinement of the theorem, as opposed to simply a special case.

Corollary 1.3. Let (X, λ) be a measure space, let $\phi_1, \ldots, \phi_{2m}$, and $\psi_1, \ldots, \psi_{2m}$ be measurable functions from X to \mathbb{C} , and assume the antisymmetric matrix

$$M_{jk} = \int_{x \in X} \phi_j(x)\psi_k(x) - \phi_k(x)\psi_j(x)\lambda(dx)$$
(1.33)

is well-defined and invertible. Then, defining

$$R(S; \phi, \psi) = \frac{1}{(m-l)! \operatorname{pf}(M)} \int_{x_{l+1}, \dots, x_m \in X} \det(\phi_j(x_k) \ \psi_j(x_k)) \prod_{l+1 \le j \le m} \lambda(dx_j), \tag{1.34}$$

we have

$$R(S; \phi, \psi) = \operatorname{pf}(K(S)), \tag{1.35}$$

where K is the antisymmetric matrix kernel

$$K(x,y) = \begin{pmatrix} \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \phi_k(y) & \sum_{1 \le j,k \le 2m} \phi_j(x) M_{jk}^{-t} \psi_k(y) \\ \sum_{1 \le j,k \le 2m} \psi_j(x) M_{jk}^{-t} \phi_k(y) & \sum_{1 \le j,k \le 2m} \psi_j(x) M_{jk}^{-t} \psi_k(y) \end{pmatrix}$$
(1.36)

Proof. Apply the previous result with $(Y, \mu) = (X, \lambda)$, $\kappa(x, y) = \delta_{xy}$, and g = 0.

In certain cases, the pfaffians simplify to determinants:

Corollary 1.4. Let (X, λ) and (Y, μ) be measure spaces, let ϕ_1, \ldots, ϕ_m be measurable functions from $X \to \mathbb{C}$, let ψ_1, \ldots, ψ_m be measurable functions from $Y \to \mathbb{C}$, and let κ be a function from $X \times Y \to \mathbb{C}$. Assume that the matrix

$$M_{jk} = \int_{x \in X, y \in Y} \phi_j(x) \kappa(x, y) \psi_k(y) \lambda(dx) \mu(dy)$$
(1.37)

is well-defined and invertible. Then, defining

$$R_{D}(S_{0}, S_{1}; \phi, \psi, \kappa) = \frac{1}{(m - l_{0})!(m - l_{1})! \det(M)} \int_{\substack{x_{l_{0}+1}, \dots x_{m} \in X \\ y_{l_{1}+1}, \dots y_{m} \in Y}} \det(\phi_{j}(x_{k})) \det(\psi_{j}(y_{k})) \det(\kappa(x_{j}, y_{k}))$$

$$\prod_{l_{0}+1 \le j \le m} \lambda(dx_{j}) \prod_{l_{1}+1 \le j \le m} \mu(dy_{j}), \quad (1.38)$$

 $we\ have$

$$R_D(S_0, S_1; \phi, \psi, \kappa) = \det \begin{pmatrix} K_{00}(S_0, S_0) & K_{01}(S_0, S_1) \\ K_{10}(S_1, S_0) & K_{11}(S_1, S_1) \end{pmatrix}, \tag{1.39}$$

where

$$K_{00}(x, x') = \sum_{1 \le j, k \le m} \phi_j(x) M_{jk}^{-t}(\kappa \cdot \psi_k)(x')$$
(1.40)

$$K_{01}(x,y) = \sum_{1 \le j,k \le m} \phi_j(x) M_{jk}^{-t} \psi_k(y)$$
(1.41)

$$K_{10}(y,x) = -\kappa(x,y) + \sum_{1 \le j,k \le m} (\kappa^t \cdot \phi_j)(y) M_{jk}^{-t}(\kappa \cdot \psi_k)(x)$$

$$\tag{1.42}$$

$$K_{11}(y, y') = \sum_{1 \le i, k \le m} (\kappa^t \cdot \phi_j)(y) M_{jk}^{-t} \psi_k(y')$$
(1.43)

for $x, x' \in X$, $y, y' \in Y$.

Corollary 1.5. Let (X, λ) be a measure space, let ϕ_1, \dots, ϕ_m , and ψ_1, \dots, ψ_m be measurable functions from X to \mathbb{C} , and assume the matrix

$$M_{jk} = \int_{x \in X} \phi_j(x)\psi_k(x)\lambda(dx)$$
 (1.44)

is well-defined and invertible. Then, defining

$$R_D(S; \phi, \psi) = \frac{1}{(m-l)! \det(M)} \int_{x_{l+1}, \dots, x_m \in X} \det(\phi_j(x_k)) \det(\psi_j(x_k)) \prod_{l+1 < j < m} \lambda(dx_j), \tag{1.45}$$

we have

$$R_D(S; \phi, \psi) = \det(K(S)), \tag{1.46}$$

where

$$K(x,y) = \sum_{1 \le i,k \le m} \phi_j(x) M_{jk}^{-t} \psi_k(y).$$
 (1.47)

2 Matrix inversions

In the cases considered below, the matrices M are principal minors of certain infinite matrices; it thus becomes crucial to determine how the inverses of the minors are related to the minors of the inverse. The key property of the matrices is that their coefficients decay as one gets farther away from the main diagonal.

We recall that a filtration on a ring R is a sequence $R = I_0 \supseteq I_1 \supseteq I_2 \dots$ of ideals of R such that $I_j I_k \subset I_{j+k}$ and $\bigcap_{1 \leq j} I_j = \{0\}$. Equivalently, a filtration can be specified by a valuation, that is a function $v : (R - \{0\}) \to \mathbb{N}$ such that

$$v(xy) \ge v(x) + v(y), \quad v(x+y) \ge \min(v(x), v(y)); \tag{2.1}$$

we simply take v(x) = j whenever I_j is the largest ideal in the filtration containing x. The ring R is complete with respect to the valuation v if R is the projective limit of the rings R/I_j ; equivalently, for any sequence $x_1, x_2, \dots \in R$ such that

$$\lim_{n \to \infty} \min_{j \neq k \ge n} v(x_j - x_k) = \infty, \tag{2.2}$$

there exists an element $x \in R$ with

$$\lim_{n \to \infty} v(x_n - x) = \infty. \tag{2.3}$$

The canonical example of a complete ring is a ring of formal power series, with valuation given by the degree map.

Given an infinite matrix M, we let M(m) denote the mth principal minor of M.

Lemma 2.1. Let R be a ring complete with respect to the valuation v, and let M be a matrix in $R^{\mathbb{Z}^+ \times \mathbb{Z}^+}$ with decaying valuations

$$v(M_{jk}) \ge |j - k| \tag{2.4}$$

and with unit diagonal elements. Then M is invertible,

$$v(M_{jk}^{-1}) \ge |j - k|,$$
 (2.5)

and for any $m \in \mathbb{Z}^+$,

$$v((M(m)^{-1} - M^{-1}(m))_{ik}) \ge 2m + 2 - j - k \tag{2.6}$$

$$v((M(m) - M^{-1}(m)^{-1})_{jk}) \ge 2m + 2 - j - k.$$
(2.7)

In particular, for j, k fixed,

$$\lim_{m \to \infty} M(m)_{jk}^{-1} = M_{jk}^{-1} \tag{2.8}$$

$$\lim_{m \to \infty} M^{-1}(m)_{jk}^{-1} = M_{jk}. \tag{2.9}$$

Proof. We first observe that for any m, $\det(M(m))$ is a unit in R; indeed, it agrees to valuation 1 with the unit product $\prod_{1 \leq j \leq m} M_{jj}$. Now, multiplication by a unit leaves the valuation unchanged, so $v(M(m)_{jk}^{-1}) = v(M(m)_{jk}^{-1} \det(M))$. This latter element is (up to sign) simply the determinant of the complementary minor to (k, j); we easily see that every term of this determinant has valuation at least m + 1 - j - k.

Now, let us consider how M(m-1) is related to $(M(m)^{-1})(m-1)^{-1}$. Recall that for a block matrix

$$M_0 = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \tag{2.10}$$

with D invertible, the upper left block of M_0^{-1} is given by $(A - BD^{-1}C)^{-1}$. In other words, the difference between the upper left block of M_0 and the inverse of the upper left block of M_0^{-1} is is $BD^{-1}C$. Applying this to M(m), we find that

$$(M(m-1) - M(m)^{-1}(m-1)^{-1})_{jk} = \frac{M(m)_{jm}M(m)_{mk}}{M(m)_{mm}};$$
(2.11)

since $M(m)_{mm}$ is a unit, we find

$$v((M(m-1)-M(m)^{-1}(m-1)^{-1})_{ik}) \ge v(M(m)_{im}) + v(M(m)_{mk}) = 2m-j-k.$$
(2.12)

By symmetry, we also find

$$v((M(m)^{-1}(m-1) - M(m-1)^{-1})_{jk}) \ge 2m - j - k.$$
(2.13)

By induction on n, we find that

$$v((M(m) - M(n)^{-1}(m)^{-1})_{jk}) \ge 2m + 2 - j - k, \tag{2.14}$$

$$v((M(n)^{-1}(m) - M(m)^{-1})_{jk}) \ge 2m + 2 - j - k.$$
(2.15)

In particular, defining an infinite matrix N by

$$N_{jk} = \lim_{n \to \infty} M(n)_{jk}^{-1}, \tag{2.16}$$

we find MN = NM = 1, and the lemma follows.

Lemma 2.2. Let R, v be as above, and let M be an infinite antisymmetric matrix such that

$$v(M_{jk}) \ge |j - k| - 1, (2.17)$$

and $M_{(2j-1)(2j)} \in \mathbb{R}^*$ for all $j \geq 1$. Then M is invertible and for all m > 0,

$$v((M(2m) - M^{-1}(2m)^{-1})_{jk}) \ge 4m + 1 - j - k.$$
(2.18)

$$v((M(2m)^{-1} - M^{-1}(2m))_{jk}) \ge \begin{cases} 2m + 2 + (j + 1 \mod 2) - k & k > j \\ 2m + 2 + (k + 1 \mod 2) - j & j > k. \end{cases}$$
(2.19)

In particular, for j, k fixed,

$$\lim_{m \to \infty} M(m)_{jk}^{-1} = M_{jk}^{-1} \tag{2.20}$$

$$\lim_{m \to \infty} M^{-1}(m)_{jk}^{-1} = M_{jk}.$$
 (2.21)

Proof. The proof is essentially as above; the main difference is that the matrix D is now 2-dimensional, of the form

$$D = \begin{pmatrix} 0 & u \\ -u & 0 \end{pmatrix},\tag{2.22}$$

for some unit u. Then, since $C = -B^t$, $(BD^{-1}C)_{jk}$ is essentially just the determinant of a 2×2 submatrix of B. For the first equation, it is trivial to determine the valuation of this determinant; for the second equation, we simply relate the determinant of a 2×2 minor of $M(2m)^{-1}$ to the determinant of the complementary minor of M(2m), and again the valuation is easy to determine.

Similarly,

Lemma 2.3. Let R, v be as above, and let M be an infinite antisymmetric matrix such that

$$v(M_{jk}) \ge |\lceil j/2 \rceil - \lceil k/2 \rceil| \tag{2.23}$$

and $M_{(2j-1)(2j)} \in \mathbb{R}^*$ for all $j \geq 1$. Then M is invertible and for all m > 0,

$$v((M(2m) - M^{-1}(2m)^{-1})_{jk}) \ge 2m + 2 - \lceil j/2 \rceil - \lceil k/2 \rceil$$
(2.24)

$$v((M(2m)^{-1} - M^{-1}(2m))_{jk}) \ge 2m + 2 - \lceil j/2 \rceil - \lceil k/2 \rceil.$$
(2.25)

In particular, for j, k fixed,

$$\lim_{m \to \infty} M(m)_{jk}^{-1} = M_{jk}^{-1} \tag{2.26}$$

$$\lim_{m \to \infty} M^{-1}(m)_{jk}^{-1} = M_{jk}.$$
 (2.27)

We digress to consider a specific matrix which arises below. For numbers α , β , we define $F(\alpha, \beta)$ to be the antisymmetric matrix with

$$F(\alpha, \beta)_{jk} = \begin{cases} \alpha^{k-j-1} \beta^{(j+1) \mod 2} \beta^{k \mod 2} & k > j \\ -\alpha^{j-k-1} \beta^{(k+1) \mod 2} \beta^{j \mod 2} & j < k. \end{cases}$$
 (2.28)

Also, if $\phi(z)$ is a Laurent series, we define the Toeplitz matrix

$$T(\phi(z))_{jk} = [z^{k-j}]\phi(z).$$
 (2.29)

The following is straightforward to verify:

Lemma 2.4. For any α , $\beta \in R$ such that $v(\alpha), v(\beta) > 0$,

$$F(\alpha, \beta) = F(-\alpha, -\beta) \tag{2.30}$$

$$F(\alpha, \beta)^{-1} = -F(-\beta, \alpha), \tag{2.31}$$

and

$$F(\alpha, 1) = T((1 - \alpha z)^{-1})F(0, 1)T((1 - \alpha z)^{-1})^{t}$$
(2.32)

$$= T((1 - \alpha/z)^{-1})F(0,1)T((1 - \alpha/z)^{-1})^t$$
(2.33)

$$F(1,\beta) = T(1+\beta z)F(1,0)T(1+\beta z)^{t}$$
(2.34)

$$= T(1 + \beta/z)F(1,0)T(1 + \beta/z)^{t}$$
(2.35)

$$F(1,0) = T((1-z^2)^{-1})F(0,1)T((1-z^2)^{-1})^t$$
(2.36)

$$= T((1-z^{-2})^{-1})F(0,1)T((1-z^{-2})^{-1})^{t}.$$
(2.37)

3 The ordinary cases: \square and \square

It will be instructive to rederive the result of [11], since this will suggest how to deal with the symmetrized cases later.

Theorem 3.1. Let p_+ , p_- be compatible parameter sets (in the sense of [12]). Then for any finite subset $S \subset \mathbb{Z}$, the probability that the set $\{\lambda_j^{\square}(p_+, p_-) - j\}$ contains S is given by

$$\det(K^{\square}(S \mid p_+, p_-)),\tag{3.1}$$

where

$$K^{\square}(a, b \mid p_+, p_-) = \sum_{1 \le l} L^{\square}(a + l \mid p_+, p_-) L^{\square}(b + l \mid p_-, p_+)$$
(3.2)

and

$$L^{\square}(a \mid p_{+}, p_{-}) = [z^{a}] \frac{E(z; p_{+})}{E(z^{-1}; p_{-})}, \tag{3.3}$$

defined by contour integration over a contour containing 0 and the zeros of $E(z^{-1}; p_{-})$ and excluding ∞ and the poles of $E(z; p_{+})$.

Proof. Since

$$\Pr(\lambda_j^{\Box}(p_+, p_-) = \lambda) = H(p_+, p_-) s_{\lambda'}(p_+) s_{\lambda'}(p_-), \tag{3.4}$$

we see that the theorem reduces formally to the symmetric function identity

$$\frac{\sum_{\lambda:S\subset\{\lambda_i-i\}} s_{\lambda'}(x)s_{\lambda'}(y)}{\sum_{\lambda} s_{\lambda'}(x)s_{\lambda'}(y)} = \det(K^{\square}(S\mid x,y)). \tag{3.5}$$

We first prove this formal identity, then consider the specific specialization of interest.

If we restrict λ so that $\ell(\lambda) \leq m$, then this only changes the left-hand-side by terms of order $O(x^m y^m)$; it will thus suffice to derive a kernel for each m such that the formal limit $m \to \infty$ of these kernels is K^{\square} .

When $\ell(\lambda) \leq m$, we find

$$s_{\lambda'}(x)s_{\lambda'}(y) = \det(e_{\lambda_k - k + j}(x))_{j,k} \det(e_{\lambda_k - k + j}(y))_{j,k}. \tag{3.6}$$

Thus we can apply Corollary 1.5 above, with

$$\phi_j(a) = e_{a+j}(x) \quad \psi_j(a) = e_{a+j}(y).$$
 (3.7)

Defining M(m) by

$$M(m)_{jk} = \sum_{a} \phi_j(a)\psi_k(a), \tag{3.8}$$

we find that M(m) is the mth principal minor of the infinite matrix

$$M_{jk} = \sum_{a} e_{a+j}(x)e_{a+k}(y) = \sum_{a} e_{a-k}(x)e_{a-j}(y), \tag{3.9}$$

for $1 \leq j, k$. Since j, k > 0, we can restrict the second sum to a > 0, and thus have

$$M = T(E(z;y))T(E(z;x))^{t}.$$
(3.10)

(Recall $T(\phi(z))_{jk} = [z^{k-j}]\phi(z)$.) We thus find

$$M^{-1} = T(E(z;x)^{-1})^{t}T(E(z;y)^{-1}), \tag{3.11}$$

With respect to the natural valuation on the ring of symmetric functions in two variables, M satisfies the hypotheses of Lemma 2.1 above; we thus find

$$\lim_{m \to \infty} (M(m)^{-1} - M^{-1}(m))_{jk} = 0 \tag{3.12}$$

for any fixed j, k. Since $v(\phi_j(a)) \ge a + j$, we find

$$\lim_{m \to \infty} \sum_{1 \le j,k \le m} \phi_j(a) M(m)_{jk}^{-t} \psi_k(a) = \sum_{1 \le j,k} \phi_j(a) M_{jk}^{-t} \psi_k(a) = \sum_{1 \le l} (\sum_{1 \le j} \phi_j(a) E(y)_{lj}^{-1}) (\sum_{1 \le k} \psi_k(a) E(x)_{lk}^{-1}). \quad (3.13)$$

We compute

$$\sum_{1 \le j} \phi_j(a) E(y)_{lj}^{-1} = \sum_{1 \le j} [z^{a+j}] E(z; x) [z^{j-l}] E(z; y)^{-1} = \sum_j [z^{a+j}] E(z; x) [z^{j-l}] E(z; y)^{-1} = [z^{a+l}] \frac{E(z; x)}{E(1/z; y)},$$
(3.14)

thus proving the desired formal result.

For any complex number u and any parameter set p, we define a specialization up on the ring of symmetric functions in x by

$$e_j(up) = u^j e_j(p). (3.15)$$

Now, specialize the formal identity by $e_j(x) \to e_j(up_+)$ and $e_j(y) \to e_j(up_-)$. For u in a neighborhood of 0, both sides converge, and thus must agree in this neighborhood. Since both sides are analytic in a neighborhood of the interval [0,1], it follows that they must agree at u=1, and the theorem is proved.

Remark 1. Since

$$\frac{E(z; p_{+})}{E(1/z; p_{-})} = \frac{H(-1/z; p_{-})}{H(-z; p_{+})},\tag{3.16}$$

we find that our operator is the same as the operator of [11] and [6] whenever the latter operator is defined.

Corollary 3.2. For any finite disjoint subsets S_+ , $S_- \subset \mathbb{Z}$, the probability that the set $\{\lambda_i^{\square}(p_+, p_-) - i\}$ contains S_+ and is disjoint from S_- is given by

$$\det \begin{pmatrix} K^{\square}(S_{+}, S_{+} \mid p_{+}, p_{-}) & \sqrt{-1}K^{\square}(S_{+}, S_{-} \mid p_{+}, p_{-}) \\ \sqrt{-1}K^{\square}(S_{+}, S_{-} \mid p_{+}, p_{-}) & I - K^{\square}(S_{-}, S_{-} \mid p_{+}, p_{-}) \end{pmatrix}$$
(3.17)

Proof. Set $T := \{\lambda_i^{\square}(p_+, p_-) - i\}$. Then the given determinant is

$$\sum_{S_0 \subset S_-} (-1)^{|S_0|} \Pr(S_+ \cup S_0 \subset T) = \Pr(S_+ \subset T, S_- \cap T = \emptyset), \tag{3.18}$$

as required. \Box

For the case \Box of signed permutations, the analogous expectation is a specialization of the symmetric function identity for \Box ; we thus have:

Corollary 3.3. Let p_+ , p_- be compatible parameter sets. Then for any finite subset $S \subset \mathbb{Z}$, the probability that the set $\{\lambda_j^{\square}(p_+, p_-) - j\}$ contains S is given by

$$\det(K^{\square}(S \mid p_+, p_-)), \tag{3.19}$$

where

$$K^{\square}(a, b \mid p_+, p_-) = \sum_{1 \le l} L^{\square}((a+l)/2 \mid p_+, p_-) L^{\square}((b+l)/2 \mid p_-, p_+), \tag{3.20}$$

defining $L^{\square}(a \mid p_+, p_-) := 0$ if $a \notin \mathbb{Z}$.

Proof. After specializing, $L^{\square}(a \mid p_+, p_-)$ becomes

$$\{[z^{a}]E(-z^{-2};p_{-})^{-1}E(-z^{2};p_{+})\} = (-1)^{a/2}L^{\square}(a/2 \mid p_{+},p_{-}).$$
(3.21)

Conjugating by $(-1)^{a/2}$ gives

$$K^{\square}(a,b \mid p_+, p_-) = \sum_{1 \le l} (-1)^{a+l} L^{\square}((a+l)/2 \mid p_+, p_-) L^{\square}((b+l)/2 \mid p_-, p_+); \tag{3.22}$$

since $L^{\square}((a+l)/2 \mid p_+, p_-) = 0$ unless a+l is even, the result follows.

Corollary 3.4. For any finite disjoint subsets S_+ , $S_- \subset \mathbb{Z}$, the probability that the set $\{\lambda_i^{\square}(p_+, p_-) - i\}$ contains S_+ and is disjoint from S_- is given by

$$\det \begin{pmatrix} K^{\square}(S_{+}, S_{+} \mid p_{+}, p_{-}) & \sqrt{-1}K^{\square}(S_{+}, S_{-} \mid p_{+}, p_{-}) \\ \sqrt{-1}K^{\square}(S_{+}, S_{-} \mid p_{+}, p_{-}) & I - K^{\square}(S_{-}, S_{-} \mid p_{+}, p_{-}) \end{pmatrix}$$
(3.23)

4 The first involution case: □

Let $\delta_{a>b}$ denote the function on $\mathbb{Z} \times \mathbb{Z}$ which is 0 when $a \leq b$ and 1 when a > b.

Theorem 4.1. Let p be a self-compatible parameter set, let α be a number with $0 \le \alpha < R(p)^{-1}$, and let p^+ be the parameter set obtained by adjoining α to r(p). Then for any finite sets $S_0, S_1 \subset \mathbb{Z}$, the probability that the set $\{\lambda_{2j-1}^{\mathbb{Z}}(p;\alpha) - 2j + 1\}$ contains S_1 and the set $\{\lambda_{2j}^{\mathbb{Z}}(p;\alpha) - 2j\}$ contains S_0 is given by

$$\operatorname{pf} \begin{pmatrix} K_{00}^{\square}(S_0, S_0 \mid p; \alpha) & K_{01}^{\square}(S_0, S_1 \mid p; \alpha) \\ K_{10}^{\square}(S_0, S_0 \mid p; \alpha) & K_{11}^{\square}(S_1, S_1 \mid p; \alpha) \end{pmatrix}$$

$$\tag{4.1}$$

where for $u, v \in \{0, 1\}$,

$$K_{uv}^{\square}(a,b|p;\alpha) = \begin{pmatrix} S_{uv}^{\square}(a,b|p;\alpha) & S_{uv}^{\square}(a,b+1|p;\alpha) \\ S_{uv}^{\square}(a+1,b|p;\alpha) & S_{uv}^{\square}(a+1,b+1|p;\alpha) \end{pmatrix} + \begin{cases} \delta_{b>a} \begin{pmatrix} \alpha^{b-a} & \alpha^{b-a+1} \\ \alpha^{b-a-1} & \alpha^{b-a} \end{pmatrix} & uv = 01 \\ -\delta_{a>b} \begin{pmatrix} \alpha^{a-b} & \alpha^{a-b-1} \\ \alpha^{a-b+1} & \alpha^{a-b} \end{pmatrix} & uv = 10 \end{cases}$$

$$(4.2)$$

with

$$S_{uv}^{\boxtimes}(a,b\mid p;\alpha) = \sum_{l>0} L_u^{\boxtimes}(a+l+1\mid p;\alpha) L_v^{\boxtimes}(b+l\mid p;\alpha) - L_u^{\boxtimes}(a+l\mid p;\alpha) L_v^{\boxtimes}(b+l+1\mid p;\alpha) \tag{4.3}$$

$$L_0^{\square}(a \mid p; \alpha) = L^{\square}(a \mid p) \tag{4.4}$$

$$L_1^{\square}(a \mid p; \alpha) = L^{\square}(a - 1 \mid p^+) \tag{4.5}$$

$$L^{\square}(a \mid p) = \delta_{a \text{ even}} - \sum_{0 < j} L^{\square}(a - 2j \mid p, p)$$

$$\tag{4.6}$$

Proof. We have

$$\Pr(\lambda^{\boxtimes}(p;\alpha) = \lambda) \propto \alpha^{f(\lambda')} s_{\lambda'}(p), \tag{4.7}$$

where $f(\lambda)$ is the number of even parts of λ , and thus

$$\alpha^{f(\lambda')} = \prod_{i} \alpha^{\lambda_{2i-1} - \lambda_{2i-2}},\tag{4.8}$$

so the result reduces to showing the corresponding symmetric function identity. And again, we may take the limit $m \to \infty$ of the kernel corresponding to the restriction $\ell(\lambda) \le 2m$.

In that case, we have

$$\alpha^{f(\lambda')} s_{\lambda'}(x) = (-1)^m \det(e_{a_k+j}(x) \ e_{b_k+j}(x)) \prod_j \alpha^{b_j - a_j - 1}, \tag{4.9}$$

with

$$a_k = \lambda_{2m-2k+2} - 2m + 2k - 2$$
 $b_k = \lambda_{2m-2k+1} - 2m + 2k - 1.$ (4.10)

Now, if we define a kernel

$$\kappa(a,b) = \alpha^{b-a-1} \delta_{b>a},\tag{4.11}$$

then for nonincreasing sequences a and b, we find

$$\det(\kappa(a_j, b_k)) = \prod_j \alpha^{b_j - a_j - 1} \tag{4.12}$$

if $a_1 < b_1 \le a_2 < b_2 \le \cdots \le a_m < b_m$; otherwise, the determinant is 0. We thus have

$$\alpha^{f(\lambda')} s_{\lambda'}(x) \propto \det(e_{a_k+j}(x) \ e_{b_k+j}(x)) \det(\kappa(a_j, b_k))$$
(4.13)

for $a_1 < a_2 < \dots a_m$ and $b_1 < b_2 < \dots b_m$. Upon symmetrizing in a and b, we can apply Corollary 1.2, with

$$\phi_j(a) = \psi_j(a) = e_{a+j}(x). \tag{4.14}$$

We have

$$(\kappa \cdot \psi_j)(a) = [z^{a+1+j}](1 - \alpha/z)^{-1} E(z; x), \tag{4.15}$$

and

$$(\kappa^t \cdot \phi_i)(a) = [z^{a-1+j}](1 - \alpha z)^{-1} E(z; x). \tag{4.16}$$

Since

$$\phi(a) + \alpha(\kappa \cdot \psi_i)(a) = (\kappa \cdot \psi_i)(a-1) \tag{4.17}$$

$$\psi(a) + \alpha(\kappa^t \cdot \phi_j)(a) = (\kappa^t \cdot \phi_j)(a+1), \tag{4.18}$$

we can simplify the matrix resulting from Corollary 1.2 by adding α times the second row/column to the first row/column and adding α times the third row/column to fourth row/column.

Now,

$$M_{jk} = \sum_{a < b} (e_{a+j}(x)e_{b+k}(x) - e_{a+k}(x)e_{b+j}(x))\alpha^{b-a-1} = \sum_{a < b} (e_{a-k}(x)e_{b-j}(x) - e_{a-j}(x)e_{b-k}(x))\alpha^{b-a-1}, \quad (4.19)$$

and thus

$$M = T(E(z;x))F(\alpha,1)T(E(z;x))^t$$

$$(4.20)$$

$$M^{-t} = T(E(1/z;x)^{-1})F(1,-\alpha)T(E(1/z;x)^{-1})^{t}$$
(4.21)

$$= T((1 - \alpha/z)E(1/z;x)^{-1}(1 - z^{-2})^{-1})F(0,1)T((1 - \alpha/z)E(1/z;x)^{-1}(1 - z^{-2})^{-1})^{t}.$$

$$(4.22)$$

Taking $v(e_j) = j$, $v(\alpha) = 1$, we see that M satisfies the hypotheses of Lemma 2.2 above. Thus if π, μ are each either of $\kappa \cdot \psi$, or $\kappa^t \cdot \phi$, we find

$$\lim_{m \to \infty} \sum_{1 \le j,k \le m} \pi_j(a) M(m)_{jk}^{-t} \mu_k(b) = \sum_{1 \le j,k} \pi_j(a) M_{jk}^{-t} \mu_k(b). \tag{4.23}$$

It thus remains to compute

$$\sum_{j>0} (\kappa \cdot \psi_j)(a) T((1-\alpha/z)E(1/z;x)^{-1}(1-z^{-2})^{-1})_{jk} = \sum_{j\geq 0} [z^{a+2j}]E(z;x)E(1/z;x)^{-1}$$
(4.24)

$$\sum_{j>0} (\kappa^t \cdot \phi_j)(a) T((1-\alpha/z)E(1/z;x)^{-1}(1-z^{-2})^{-1})_{jk} = \sum_{j\geq 0} [z^{a+2j}](1-\alpha/z)E(z;x)E(1/z;x)^{-1}(1-\alpha z)^{-1}.$$
(4.25)

This gives the theorem, once we observe that

$$\sum_{j} [z^{a+2j}] E(z;x) E(1/z;x)^{-1} = \delta_{a \text{ even}}.$$
 (4.26)

Remark 1. The fact that K_{00} is independent of α corresponds to the fact that the joint distribution of the even rows of $\lambda^{\square}(p;\alpha)$ is independent of α , as remarked in Section 7 of [3]. Similarly, the structure of K_{11} corresponds to the fact that the odd rows of $\lambda^{\square}(p;\alpha)$ are distributed as the odd rows of $\lambda^{\square}(p^+;0)$ (which are equal to the even rows).

Remark 2. The point of using

$$L^{\square}(a \mid p) = \delta_{a \text{ even}} - \sum_{0 < j} L^{\square}(a - 2j \mid p, p)$$

$$\tag{4.27}$$

instead of

$$L^{\square}(a \mid p) = \sum_{j>0} L^{\square}(a+2j \mid p, p)$$
 (4.28)

is that the latter only converges for p^+ when $\alpha \leq 1$ (and converges to an incorrect value for $\alpha = 1$).

Remark 3. We observe the following relation between L_0^{\square} and L_1^{\square} :

$$\alpha L_0^{\square}(a+1\mid p;\alpha) - L_0^{\square}(a\mid p;\alpha) = \alpha L_1^{\square}(a\mid p;\alpha) - L_1^{\square}(a+1\mid p;\alpha). \tag{4.29}$$

Corollary 4.2. With hypotheses as above, and $\alpha = 1$, the conclusion holds with

$$S_{00}^{\boxtimes}(a, b \mid p; 1) = S_{00}^{\boxtimes}(a, b \mid p) \tag{4.30}$$

$$S_{01}^{\square}(a,b\mid p;1) = -L^{\square}(a+1\mid p) - S_{00}^{\square}(a,b\mid p) \tag{4.31}$$

$$S_{10}^{\square}(a, b \mid p; 1) = L^{\square}(b + 1 \mid p) - S_{00}^{\square}(a, b \mid p)$$

$$\tag{4.32}$$

$$S_{11}^{\square}(a,b \mid p;1) = L^{\square}(a+1 \mid p) - L^{\square}(b+1 \mid p) + S_{00}^{\square}(a,b \mid p)$$

$$(4.33)$$

Proof. We compute

$$\frac{E(z;p^+)}{E(1/z;p^+)} = \frac{-E(z;p)}{zE(1/z;p)},\tag{4.34}$$

so

$$L_1^{\square}(a \mid p; 1) = \delta_{a \text{ odd}} + \sum_{0 < j} [z^{a-2j}] \frac{E(z; p)}{E(1/z; p)}$$
(4.35)

$$=1-L^{\boxtimes}(a\mid p). \tag{4.36}$$

If we do not wish to separate the odd and even rows, we have:

Corollary 4.3. Let p be a self-compatible parameter set, let α be a number with $0 \le \alpha < R(p)^{-1}$, and let p^+ be the parameter set obtained by adjoining α to r(p). Then for any finite subset $S \subset \mathbb{Z}$, the probability that $\{\lambda_j^{\mathbb{Z}}(p;\alpha) - j\}$ contains S is given by

$$pf(K^{\boxtimes'}(S \mid p; \alpha)), \tag{4.37}$$

with

$$K^{\boxtimes'}(\mid p; \alpha) = \begin{pmatrix} S_{00}^{\boxtimes'}(\mid p; \alpha) & S_{01}^{\boxtimes'}(\mid p; \alpha) \\ S_{10}^{\boxtimes'}(\mid p; \alpha) & S_{11}^{\boxtimes'}(\mid p; \alpha) - \epsilon^{\boxtimes}(\mid \alpha) \end{pmatrix}$$
(4.38)

$$S_{uv}^{\boxtimes'}(a,b\mid p;\alpha) = \sum_{l>0} L_u^{\boxtimes'}(a+l+1\mid p;\alpha) L_v^{\boxtimes'}(b+l\mid p;\alpha) - L_u^{\boxtimes'}(a+l\mid p;\alpha) L_v^{\boxtimes'}(b+l+1\mid p;\alpha) \tag{4.39}$$

$$L_0^{\boxtimes'}(a \mid p; \alpha) = (-\alpha)^{a \bmod 2} - \sum_{0 < j} L^{\square}(a - 2j \mid p, p^+)$$
(4.40)

$$L_1^{\boxtimes'}(a \mid p; \alpha) = -L^{\square}(a - 1 \mid p^+, p)$$
(4.41)

$$\epsilon^{\square}(a,b\mid\alpha) = \alpha^{|b-a|-1}\operatorname{sgn}(b-a). \tag{4.42}$$

Proof. The key step is to sum over the subsets of S. By the theorem, we have

$$\sum_{S' \subset S} \Pr(S \subset \{\lambda_j - j\}) = \sum_{S_0, S_1 \subset S} \operatorname{pf} \begin{pmatrix} K_{00}^{\square}(S_0, S_0) & K_{01}^{\square}(S_0, S_1) \\ K_{10}^{\square}(S_1, S_0) & K_{11}^{\square}(S_0, S_1) \end{pmatrix}$$
(4.43)

$$= \operatorname{pf} \left(J + \begin{pmatrix} K_{00}^{\square}(S, S) & K_{01}^{\square}(S, S) \\ K_{10}^{\square}(S, S) & K_{11}^{\square}(S, S) \end{pmatrix} \right), \tag{4.44}$$

where J is the kernel

$$J(a,b) = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}. \tag{4.45}$$

Subtract α times the second and third rows from the first and fourth rows (respectively), then subtract the first row from the fourth and the third from the second, then apply the same transformations to the columns. This transformation is symplectic (preserves J), and forces the last row of the K matrix to 0. We may thus expand along the bottom row, giving

$$\sum_{S' \subset S} \Pr(S \subset \{\lambda_j - j\}) = \operatorname{pf}(J + K^{\boxtimes'}(S, S)) = \sum_{S' \subset S} \operatorname{pf}(K^{\boxtimes'}(S')), \tag{4.46}$$

since

$$L_0^{\boxtimes'}(a \mid p; \alpha) = L_0^{\boxtimes}(a \mid p; \alpha) - \alpha L_0^{\boxtimes}(a + 1 \mid p; \alpha)$$

$$\tag{4.47}$$

$$L_1^{\boxtimes'}(a \mid p; \alpha) = L_0^{\boxtimes}(a+1 \mid p; \alpha) - L_1^{\boxtimes}(a \mid p; \alpha). \tag{4.48}$$

Thus

$$\Pr(S \subset \{\lambda_j - j\}) = \operatorname{pf}(K^{\boxtimes'}(S')) \tag{4.49}$$

as required. \Box

Remark. We could also have proved this directly via Theorem 1.1 above, with $\phi_j(a) = e_{a+j}(x)$ and $\epsilon(a,b) = \epsilon^{\square}(a,b)$.

Corollary 4.4. For any finite disjoint subsets S_+ , $S_- \subset \mathbb{Z}$, the probability that $\{\lambda_i^{\mathbb{Z}}(p;\alpha) - i\}$ contains S_+ and is disjoint from S_- is

$$\operatorname{pf}\begin{pmatrix} K^{\boxtimes'}(S_{+}, S_{+} \mid p; \alpha) & \sqrt{-1}K^{\boxtimes'}(S_{+}, S_{-} \mid p; \alpha) \\ \sqrt{-1}K^{\boxtimes'}(S_{-}, S_{+} \mid p; \alpha) & J - K^{\boxtimes'}(S_{-}, S_{-} \mid p; \alpha) \end{pmatrix}$$
(4.50)

5 The second involution case: \square

Similarly, for the other involution case, we have

Theorem 5.1. Let p be a self-compatible parameter set, let β be a number with $0 \le \beta < Q(p)^{-1}$, and let p^+ be the parameter set obtained by adjoining β to q(p). Then for any finite sets $S_0, S_1 \subset \mathbb{Z}$, the probability that the set $\{\lambda_{2j-1}^{\mathbb{Z}}(p;\beta)-2j+1\}$ contains S_1 and the set $\{\lambda_{2j}^{\mathbb{Z}}(p;\beta)-2j\}$ contains S_0 is given by

$$\operatorname{pf}\begin{pmatrix} K_{00}^{\mathbb{N}}(S_{0}, S_{0} \mid p; \beta) & K_{01}^{\mathbb{N}}(S_{0}, S_{1} \mid p; \beta) \\ K_{10}^{\mathbb{N}}(S_{1}, S_{0} \mid p; \beta) & K_{11}^{\mathbb{N}}(S_{1}, S_{1} \mid p; \beta) \end{pmatrix},$$
(5.1)

where

$$K_{00}^{\square}(a, b \mid p; \beta) = \begin{pmatrix} S_{00}^{\square}(a, b \mid p; \beta) & S_{01}^{\square}(a, b \mid p; \beta) \\ S_{10}^{\square}(a, b \mid p; \beta) & S_{11}^{\square}(a, b \mid p; \beta) \end{pmatrix}$$
(5.2)

$$K_{01}^{\square}(a,b \mid p;\beta) = \begin{pmatrix} S_{02}^{\square}(a,b \mid p;\beta) & S_{00}^{\square}(a,b \mid p;\beta) \\ S_{12}^{\square}(a,b \mid p;\beta) & S_{10}^{\square}(a,b \mid p;\beta) \end{pmatrix} + \delta_{b>a} \begin{pmatrix} 0 & 0 \\ \beta^{a \operatorname{mod} 2}\beta^{(b+1)\operatorname{mod} 2} & 0 \end{pmatrix}$$

$$K_{10}^{\square}(a,b \mid p;\beta) = \begin{pmatrix} S_{20}^{\square}(a,b \mid p;\beta) & S_{21}^{\square}(a,b \mid p;\beta) \\ S_{00}^{\square}(a,b \mid p;\beta) & S_{01}^{\square}(a,b \mid p;\beta) \end{pmatrix} - \delta_{a>b} \begin{pmatrix} 0 & \beta^{(a+1)\operatorname{mod} 2}\beta^{b \operatorname{mod} 2} \\ 0 & 0 \end{pmatrix}$$
(5.4)

$$K_{10}^{\square}(a, b \mid p; \beta) = \begin{pmatrix} S_{20}^{\square}(a, b \mid p; \beta) & S_{21}^{\square}(a, b \mid p; \beta) \\ S_{00}^{\square}(a, b \mid p; \beta) & S_{01}^{\square}(a, b \mid p; \beta) \end{pmatrix} - \delta_{a > b} \begin{pmatrix} 0 & \beta^{(a+1) \bmod 2} \beta^{b \bmod 2} \\ 0 & 0 \end{pmatrix}$$
 (5.4)

$$K_{11}^{\square}(a,b \mid p;\beta) = \begin{pmatrix} S_{22}^{\square}(a,b \mid p;\beta) & S_{20}^{\square}(a,b \mid p;\beta) \\ S_{02}^{\square}(a,b \mid p;\beta) & S_{00}^{\square}(a,b \mid p;\beta) \end{pmatrix}$$

$$(5.5)$$

$$S_{uv}^{\square}(a,b \mid p;\beta) = \sum_{l>0} L_u^{\square}(a+l+1 \mid p;\beta) L_v^{\square}(b+l \mid p;\beta) - L_u^{\square}(a+l \mid p;\beta) L_v^{\square}(b+l+1 \mid p;\beta)$$
 (5.6)

$$L_0^{\mathbb{N}}(a \mid p; \beta) = L^{\square}(a \mid p, p^+)$$
 (5.7)

$$L_1^{\square}(a \mid p; \beta) = \begin{cases} \sum_{j \ge 0} L^{\square}(a + 2j + 1 \mid p, p) & a \text{ even} \\ \beta \sum_{j \ge 0} L^{\square}(a + 2j + 2 \mid p^+, p^+) & a \text{ odd} \end{cases}$$
 (5.8)

$$L_{2}^{\square}(a \mid p; \beta) = \begin{cases} \beta - \beta \sum_{j \geq 0} L^{\square}(a + 2j \mid p, p) & a \text{ even} \\ 1 - \sum_{j \geq 0} L^{\square}(a + 2j + 1 \mid p^{+}, p^{+}) & a \text{ odd} \end{cases}$$
 (5.9)

Proof. As above, we reduce to an application of Corollary 1.2, with

$$\phi_i(a) = \psi_i(a) = e_{a+i}(x) \tag{5.10}$$

and

$$\kappa(a,b) = \delta_{b>a} \beta^{a \bmod 2} \beta^{(b+1) \bmod 2}. \tag{5.11}$$

We compute

$$(\kappa \cdot \psi_j)(a) = \begin{cases} [z^{a+1+j}](1+\beta/z)E(z;x)(1-1/z^2)^{-1} & a \text{ even} \\ [z^{a+2+j}]\beta(1+\beta z)E(z;x)(1-1/z^2)^{-1} & a \text{ odd} \end{cases}$$
(5.12)

$$(\kappa^{t} \cdot \phi_{j})(a) = \begin{cases} [z^{a+j}]\beta(1+\beta/z)E(z;x)z^{2}(1-z^{2})^{-1} & a \text{ even} \\ [z^{a+1+j}](1+\beta z)E(z;x)z^{2}(1-z^{2})^{-1} & a \text{ odd} \end{cases}$$
(5.13)

and

$$M_{jk} = \sum_{b>a} (e_{a+j}(x)e_{b+k}(x) - e_{b+j}(x)e_{a+k}(x))\beta^{a \bmod 2}\beta^{(b+1) \bmod 2}.$$
 (5.14)

Now, when $j \mod 2 \neq k \mod 2$, we can simply shift the variables of summation to obtain

$$M_{jk} = (T(E(z;x))F(1,\beta)T(E(z;x))^t)_{jk}.$$
(5.15)

When $j \mod 2 = k \mod 2$, this gives

$$M_{jk} = \sum_{b>a} (e_{a-k}(x)e_{b-j}(x) - e_{b-k}(x)e_{a-j}(x))\beta^{a \bmod 2}\beta^{(b+1) \bmod 2}$$
(5.16)

$$= \sum_{a} \beta^{a \operatorname{mod} 2} e_{a-j}(x) \sum_{b>a} (e_{b-2j+k}(x) - e_{b-k}(x)) \beta^{(b+1)\operatorname{mod} 2}$$
(5.17)

$$= \sum_{a} \beta^{a \operatorname{mod} 2} e_{a-j}(x) \sum_{b \le a} (e_{b-k}(x) - e_{b-2j+k}(x)) \beta^{(b+1) \operatorname{mod} 2}$$
(5.18)

$$= \sum_{b \le a} (e_{a-j}(x)e_{b-k}(x) - e_{a-k}(x)e_{b-j}(x))\beta^{a \bmod 2}\beta^{(b+1) \bmod 2},$$
(5.19)

so we conclude that

$$M = T(E(z;x))F(1,\beta)T(E(z;x))^{t}$$
(5.20)

$$M^{-t} = T(E(1/z;x)^{-1})F(-\beta,1)T(E(1/z;x)^{-1})^{t}$$
(5.21)

$$= T(E(1/z;x)^{-1}(1+\beta/z)^{-1})F(0,1)T(E(1/z;x)^{-1}(1+\beta/z)^{-1})^{t}.$$
 (5.22)

In particular, M^{-1} satisfies the hypotheses of Lemma 2.2, so the kernels for finite m tend to a limit. We thus readily compute the kernel given above.

Corollary 5.2. Let p be a self-compatible parameter set, let β be a number with $0 \leq \beta < Q(p)^{-1}$, and let p^+ be the parameter set obtained by adjoining β to q(p). Then for any finite subset $S \subset \mathbb{Z}$, the probability that $\{\lambda_j^{\mathbb{Z}}(p;\beta) - j\}$ contains S is given by

$$pf(K^{\square'}(S \mid p; \beta)), \tag{5.23}$$

with

$$K^{\boxtimes'}(\mid p; \beta) = \begin{pmatrix} S_{00}^{\boxtimes'}(\mid p; \beta) & S_{01}^{\boxtimes'}(\mid p; \beta) \\ S_{10}^{\boxtimes'}(\mid p; \beta) & S_{11}^{\boxtimes'}(\mid p; \beta) - \epsilon^{\boxtimes}(\mid \beta) \end{pmatrix}$$
(5.24)

$$S_{uv}^{\boxtimes'}(a,b\mid p;\beta) = \sum_{l>0} L_u^{\boxtimes'}(a+l+1\mid p;\beta) L_v^{\boxtimes'}(b+l\mid p;\beta) - L_u^{\boxtimes'}(a+l\mid p;\beta) L_v^{\boxtimes'}(b+l+1\mid p;\beta) \tag{5.25}$$

$$L_0^{\boxtimes'}(a \mid p; \beta) = L^{\square}(a \mid p, p^+)$$
 (5.26)

$$L_1^{\mathbb{N}'}(a \mid p; \beta) = -\beta^{(a+1) \bmod 2} + \sum_{j \ge 0} L^{\square}(a+2j+1 \mid p^+, p)$$
(5.27)

$$\epsilon^{\square}(a, b \mid \beta) = \beta^{(\max(a, b) + 1) \bmod 2} \beta^{\min(a, b) \bmod 2} \operatorname{sgn}(b - a). \tag{5.28}$$

For the case \boxtimes of hyperoctahedral involutions, similar arguments can be used to derive the kernel for general α and β . Since this is rather complicated, we consider only the distribution of $\{\lfloor \lambda_{2j-1}/2 \rfloor - j\}$ and $\{\lfloor \lambda_{2j}/2 \rfloor - j\}$; or equivalently, the distribution for $\beta = 0$.

Theorem 6.1. Let p be a self-compatible parameter set, let α be a number with $0 \le \alpha < R(p)^{-1}$, let β be a number with $0 \le \beta < Q(p)^{-1}$, and let p^+ be the parameter set obtained by adjoining α to r(p). Then for any finite subsets $S_0, S_1 \subset \mathbb{Z}$, the probability that $\{\lfloor \lambda_{2j-1}^{\boxtimes}(p;\alpha,\beta)/2 \rfloor - j\}$ contains S_1 and $\{\lfloor \lambda_{2j}^{\boxtimes}(p;\alpha,\beta)/2 \rfloor - j\}$ contains S_0 is given by

$$\det \begin{pmatrix} K_{00}^{\boxtimes}(S_0, S_0 \mid p; \alpha) & K_{01}^{\boxtimes}(S_1, S_0 \mid p; \alpha) \\ K_{10}^{\boxtimes}(S_0, S_1 \mid p; \alpha) & K_{11}^{\boxtimes}(S_1, S_1 \mid p; \alpha) \end{pmatrix}, \tag{6.1}$$

where

$$K_{00}^{\boxtimes}(a, b \mid p; \alpha) = \sum_{l>0} L^{\square}(a + l \mid p, p) L^{\square}(b + l \mid p, p)$$
(6.2)

$$K_{01}^{\boxtimes}(a, b \mid p; \alpha) = \sum_{l>0} L^{\square}(a + l \mid p, p) L^{\square}(b + l \mid p, p^{+})$$
(6.3)

$$K_{10}^{\boxtimes}(a, b \mid p; \alpha) = \sum_{l>0} L^{\square}(a + l \mid p^+, p) L^{\square}(b + l \mid p, p) - \delta_{a \ge b} \alpha^{a-b}$$
(6.4)

$$K_{11}^{\boxtimes}(a, b \mid p; \alpha) = \sum_{l>0} L^{\square}(a + l \mid p^+, p) L^{\square}(b + l \mid p, p^+)$$
(6.5)

Proof. We apply Corollary 1.4, with

$$\phi_j(a) = \psi_j(a) = e_{a+j}(x),$$
(6.6)

and

$$\kappa(a,b) = \delta_{b>a} \alpha^{b-a}. \tag{6.7}$$

We find

$$M = T(E(z;x))T(E(z;x)/(1-\alpha z))^{t}$$
(6.8)

$$M^{-t} = T(E(z;x)^{-1})^{t} T(E(z;x)^{-1} (1 - \alpha z))$$
(6.9)

The theorem follows immediately.

Remark. For general β , we instead apply Corollary 1.2, with

$$\phi_j(a) = \psi_j(a) = e_{(a+j)/2}(x) \tag{6.10}$$

(using the convention that $e_{a/2}(x) = 0$ if a is odd) and

$$\kappa(a,b) = \delta_{b>a} \sqrt{\alpha}^{b-a-1} \sqrt{-\beta}^{a \operatorname{mod} 2} \sqrt{-\beta}^{(b+1) \operatorname{mod} 2}.$$
(6.11)

We then have

$$M = T(E(z^2; x))F(\sqrt{\alpha}, \sqrt{-\beta})T(E(z^2; x))^t$$
(6.12)

$$M^{-t} = T(E(z^2; x)^{-1})^t F(-\sqrt{-\beta}, \sqrt{\alpha}) T(E(z^2; x)^{-1})$$
(6.13)

and M satisfies the hypotheses of Lemma 2.3 above. The details are left to the interested reader. (The individual terms of the resulting operator are all fairly simple; however, since the operator depends strongly on the parity of a and b, there are a total of 10 such terms to consider.)

7 Other identities

There are three Littlewood identities that were not considered in [3]:

$$\sum_{\lambda=(\alpha+1|\alpha)} s_{\lambda'}(x) = \prod_{j < k} (1 + x_j x_k) \tag{7.1}$$

$$\sum_{\lambda = (\alpha - 1 \mid \alpha)} s_{\lambda'}(x) = \prod_{j} (1 + x_j^2) \prod_{j < k} (1 + x_j x_k)$$
(7.2)

$$\sum_{\lambda = (\alpha - 1|\alpha)} s_{\lambda'}(x) = \prod_{j} (1 + x_j^2) \prod_{j < k} (1 + x_j x_k)$$

$$\sum_{\lambda = (\alpha|\alpha)} (-1)^{(|\lambda| - p(\lambda))/2} s_{\lambda'}(x) = \prod_{j} (1 + x_j) \prod_{j < k} (1 - x_j x_k),$$
(7.2)

where $(\alpha|\beta)$ is Frobenius notation, and $p((\alpha|\beta))$ is equal to the number of parts of α . We also note the following special case of the third identity:

$$\sum_{\lambda=(\alpha|\alpha)} \tilde{s}_{\lambda'}(x) = \prod_{j,k} (1 + x_j x_k) \tag{7.4}$$

For the first, second, and fourth identity, there exists an explicit combinatorial correspondence proving the identity; in the first two cases, this is given by [7], while the third case simply corresponds to increasing subsequences of multisets with rotational symmetry by 90 degrees. These correspondences extend to the case of an arbitrary parameter set p such that p is compatible with its conjugate p'.

As remarked in [9], these identities can be shown via the Cauchy-Binet theorem. But then Corollary 1.5 implies that the corresponding correlation functions are given in principle by appropriate determinants.

For instance,

Theorem 7.1. For any parameter set p compatible with its conjugate and any finite subset $S \subset \mathbb{Z}$,

$$\frac{\sum_{\lambda=(\alpha-1|\alpha)} s_{\lambda'}(p)}{\sum_{\lambda=(\alpha-1|\alpha)} s_{\lambda'}(p)} = \det(K(S)), \tag{7.5}$$

where

$$K(a,b) = (-1)^{(|b|-b)/2} \sum_{l} (-1)^{(|l|-l)/2} L^{\square}(a+|l| \mid p,p') L^{\square}(|b|+l \mid p',p).$$
 (7.6)

Remark 1. We use $\{\lambda_i - i + 1\}$ instead of $\{\lambda_i - i\}$ in order to increase symmetry. In particular, note that λ is of the appropriate form if and only if the set $\{\lambda_i - i + 1\}$ contains precisely one element of $\{j, -j\}$ for each j.

Remark 2. As written, the kernel is only explicitly defined for sufficiently small parameter sets, and must be analytically continued to the general case.

Proof. For simplicity, we consider instead

$$\sum_{\substack{\lambda = (\alpha - 1 \mid \alpha) \\ S \subset \{\lambda_i - i + 1\}}} (-1)^{|\lambda|/2} s_{\lambda'}(p), \tag{7.7}$$

which naturally differs only by rescaling p by $\sqrt{-1}$.

We find that for λ of the appropriate form with $\ell(\lambda) \leq m$,

$$(-1)^{|\lambda|/2} s_{\lambda'}(p) = \det(\phi_j(a_k)) \det(\psi_j(a_k))_{0 \le j \le m}, \tag{7.8}$$

where

$$\phi_i(a) = e_{i+a}(p) \tag{7.9}$$

$$\psi_j(a) = \delta_{|a|=j} \tag{7.10}$$

and $a_k = \lambda_{m+1-i} - m + i$. We then apply Corollary 1.5, with

$$M_{jk} = \begin{cases} e_j & k = 0 \\ e_{j+k} + e_{j-k} & k > 0 \end{cases}; \tag{7.11}$$

in particular M satisfies the hypotheses of Lemma 2.1. We readily verify that

$$M_{jk}^{-1} = (-1)^j \sum_{l} (-1)^l [w^{j-l} z^{k-|l|}] H(1/w; p) H(w; p) E(z; p), \tag{7.12}$$

so

$$\sum_{j,k>0} \phi_j(a) M_{jk}^{-t} \psi_j(b) = \sum_{j>0} e_{j+a}(p) M_{|b|j}^{-1}$$
(7.13)

$$= (-1)^{|b|} \sum_{l} (-1)^{l} [w^{|b|-l} z^{-a-|l|}] H(1/w; p) H(w; p) E(z; p) E(1/z; p)$$
 (7.14)

$$= \sum_{l} [w^{|b|-l} z^{a+|l|}] \frac{E(z;p)E(1/z;p)}{E(w;p)E(1/w;p)}. \tag{7.15}$$

Scaling p by $\sqrt{-1}$ and simplifying gives the desired result.

Dually,

Corollary 7.2. For any parameter set p compatible with its conjugate and any finite subset $S \subset \mathbb{Z}$,

$$\frac{\sum_{\lambda=(\alpha+1|\alpha)} s_{\lambda'}(p)}{\sum_{\lambda=(\alpha+1|\alpha)} s_{\lambda'}(p)} = \det(I - K(S)), \tag{7.16}$$

where

$$K(a,b) = (-1)^{(|b|+b)/2} \sum_{l} (-1)^{(|l|-l)/2} L^{\square}(-a+|l| \mid p,p') L^{\square}(|b|+l \mid p',p).$$

$$(7.17)$$

For the remaining Littlewood identity, we similarly have:

Theorem 7.3. For any parameter set p compatible with its conjugate and any finite subset $S \subset \mathbb{Z} + 1/2$,

$$\frac{\sum_{\substack{\lambda=(\alpha|\alpha)\\S\subset\{\lambda_i-i+1/2\}}} (-1)^{(|\lambda|+p(\lambda))/2} s_{\lambda'}(p)}{\sum_{\substack{\lambda=(\alpha|\alpha)}} (-1)^{(|\lambda|+p(\lambda))/2} s_{\lambda'}(p)} = \det(K(S)), \tag{7.18}$$

where

$$K(a,b) = \sum_{l \in \mathbb{Z} + 1/2} [z^{a+|l|} w^{|b|-l}] \frac{E(z;p)E(1/z;p)}{E(w;p)E(1/w;p)}.$$
 (7.19)

Proof. We take

$$\phi_j(a) = e_{j+a+1/2}(p) \tag{7.20}$$

$$\psi_j(a) = \delta_{|a|=j+1/2},$$
(7.21)

so

$$M_{jk} = e_{j+k+1}(p) + e_{j-k}(p) (7.22)$$

We find

$$M_{jk}^{-1} = (-1)^j \sum_{l} (-1)^l [t^{j-l} u^{k+1/2-|l+1/2|}] H(1/t; p) H(t; p) E(u; p), \tag{7.23}$$

and thus obtain the stated kernel.

Specializing, we obtain (for an appropriate definition of $\lambda^{\circ}(p)$, corresponding to increasing subsequences of multisets with rotational symmetry):

Corollary 7.4. Let p be a parameter set compatible with its conjugate. Then for any finite subset $S \subset \mathbb{Z} + 1/2$,

$$\Pr(S \subset \{\lambda_i^{\circ}(p) - i + 1/2\}) = \det(K(S)), \tag{7.24}$$

where

$$K(a,b) = \sum_{l \in \mathbb{Z}+1/2} \left[z^{a+|l|} w^{|b|-l} \right] \frac{E(\sqrt{-1}z^2; p) E(\sqrt{-1}/z^2; p)}{E(\sqrt{-1}w^2; p) E(\sqrt{-1}/w^2; p)}.$$
(7.25)

8 Fredholm pfaffians

Let J be the kernel

$$J(a,b) = \delta_{ab} \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \tag{8.1}$$

Then for any other antisymmetric kernel K, we have

$$pf((J+K)(S)) = \sum_{S' \subset S} pf(K(S')).$$
(8.2)

This suggests the correct way to extend to the infinite case, thus generalizing Fredholm determinants. We define the Fredholm pfaffian

$$pf(J+K)_X := \int_{S \subset X} pf(K(S))\lambda(dS), \tag{8.3}$$

where $\lambda(dS)$ is the natural induced measure on the space of finite subsets of X; by convention, $\lambda(\{\emptyset\}) = 1$. In particular, when X is finite and λ is the counting measure, we have

$$pf(J+K)_X = \sum_{S \subset X} pf(K(S)) = pf(J+K), \tag{8.4}$$

as we would expect. Naturally, this includes Fredholm determinants as special cases, since

$$pf(J + \begin{pmatrix} \epsilon & K \\ -K & 0 \end{pmatrix}) = det(I + K), \tag{8.5}$$

for any scalar kernel K and any antisymmetric scalar kernel ϵ .

We note the following properties of Fredholm pfaffians:

Lemma 8.1. For any antisymmetric matrix kernel K,

$$pf(J+K)_X^2 = \det(I+J^{-1}K)_X.$$
(8.6)

For any ordinary matrix kernel K_0 ,

$$pf((I+K_0)(J+K)(I+K_0^t))_X = \det(I+K_0)_X pf(J+K)_X.$$
(8.7)

If A is a matrix operator from X to Y, M_X is an invertible antisymmetric matrix operator on X, and M_Y is an invertible antisymmetric matrix operator on Y, then

$$pf(M_Y)_Y pf(M_Y^{-t} + AM_X A^t)_Y = pf(M_X)_X pf(M_X^{-t} + A^t M_Y A)_X.$$
(8.8)

Remark. The last equation generalizes the Fredholm determinant identity

$$\det(M_1)\det(M_1^{-1} + AM_2B) = \det(M_2)\det(M_2^{-1} + BM_1A). \tag{8.9}$$

The significance of Fredholm pfaffians for our purposes is related to the following result:

Theorem 8.2. Let (X, λ) be a measure space, and let μ be a measure on the set of countable subsets of X. Suppose

$$\int_{T \subset X} \chi_T(dS)\mu(dT) = \operatorname{pf}(K(S))\lambda(dS), \tag{8.10}$$

where χ_T is the atomic measure concentrated on the finite subsets of T. Then for functions $f: X \to \mathbb{C}$,

$$\int_{T \subset X} \prod_{x \in T} (1 + f(x))\mu(dT) = \operatorname{pf}(J + \sqrt{f}K\sqrt{f})_{X,\lambda}$$
(8.11)

whenever both sides are defined.

Proof. On the one hand, we have

$$\int_{T \subset X} \prod_{x \in T} (1 + f(x))\mu(dT) = \int_{T \subset X} \sum_{S \subset T} \prod_{x \in S} f(x)\mu(dT) = \int_{T \subset X} \int_{S \subset X} \prod_{x \in S} f(x)\chi_T(dS)\mu(dT); \tag{8.12}$$

on the other hand, we have

$$pf(J + \sqrt{f}K\sqrt{f})_{X,\lambda} = \int_{S \subset X} pf(\sqrt{f}K\sqrt{f})(S)\lambda(dS) = \int_{S \subset X} \prod_{x \in S} f(x) pf(K(S))\lambda(dS). \tag{8.13}$$

The theorem follows. \Box

Remark 1. Note that

$$pf(J + \sqrt{f}K\sqrt{f})_{X,\lambda} = pf(J + K)_{X,f\lambda}, \tag{8.14}$$

where $(f\lambda)(dx) = f(x)\lambda(dx)$; thus the square root is best thought of as merely notational.

Note in particular that if $X = \mathbb{Z}$, λ is the counting measure, and μ is a probability measure, then $\mathbf{E}(\chi_T(\{S\}))$ is precisely equal to $\Pr(S \subset T)$, thus explaining the connection with our earlier results.

In particular, Theorem 1.1 is related to a Fredholm pfaffian result:

Theorem 8.3. Let (X, λ) be a measure space, let $f, \phi_1, \dots, \phi_{2m}$, be functions from X to \mathbb{C} , let ϵ be an antisymmetric function from $X \times X$ to \mathbb{C} , and assume the antisymmetric matrix

$$M_{jk} = \int_{x,y \in X} \phi_j(x)\epsilon(x,y)\phi_k(y)\lambda(dx)\lambda(dy)$$
(8.15)

is well-defined and invertible. Then

$$F(f;\phi,\epsilon) := \frac{1}{(2m)! \operatorname{pf}(M)} \int_{x_1,\dots,x_{2m} \in X} \det(\phi_j(x_k)) \operatorname{pf}(\epsilon(x_j, x_k)) \prod_{1 \le j \le 2m} (1 + f(x_j)) \lambda(dx_j)$$
(8.16)

$$= \operatorname{pf}(J + \sqrt{f}K\sqrt{f})_{X,\lambda}, \tag{8.17}$$

in the sense that if either side is defined, then both are defined, and take the same value.

Proof. This of course follows immediately from Theorem 1.1, but the following independent proof (based on the arguments of [13]) gives useful insight into how the kernel K can be derived. (The above proof, of course, has the advantage of using only finite methods.)

From Section 4 of [8], we have

$$\int_{x_1, x_2, \dots x_{2m}} \det(\phi_j(x_k)) \operatorname{pf}(\epsilon(x_j, x_k)) \prod_j \mu(dx_j) = (2m)! \operatorname{pf}(\int_{x, y} \phi_j(x) \epsilon(x, y) \phi_k(y) \mu(dx) \mu(dy))$$
(8.18)

for any measure μ . Thus, taking $\mu = (1+f)\lambda$, we find

$$F(f;\phi,\epsilon) = \operatorname{pf}(M)^{-1} \operatorname{pf}\left(\int_{x,y\in X} \phi_j(x)\epsilon(x,y)\phi_k(y)(1+f(x))(1+f(y))\lambda(dx)\lambda(dy)\right)$$
(8.19)

$$= pf(M)^{-1} pf(M + AM_X A^t)$$
(8.20)

$$= pf(M_X)_X pf(M_X^{-t} + A^t M^{-t} A)_X$$
(8.21)

where

$$A = \begin{pmatrix} \sqrt{f}\phi_j & \sqrt{f}\epsilon \cdot \phi_j \end{pmatrix} \qquad M_X = \begin{pmatrix} \sqrt{f}(x)\epsilon(x,y)\sqrt{f}(y) & I \\ -I & 0 \end{pmatrix}$$
(8.22)

We thus find $pf(M_X)_X = 1$ and

$$M_X^{-t} = \begin{pmatrix} 0 & I \\ -I & -\sqrt{f(x)}\epsilon(x,y)\sqrt{f(y)} \end{pmatrix}. \tag{8.23}$$

Thus

$$F(f;\phi,\epsilon) = \operatorname{pf}(J + \sqrt{f}K\sqrt{f})_X \tag{8.24}$$

as required. \Box

Let λ be a random partition. We say that the distribution of λ is represented by the antisymmetric kernel K(a,b) on $\mathbb Z$ if

$$\Pr(S \subset \{\lambda_i - i\}) = \operatorname{pf}(K(S)). \tag{8.25}$$

(Thus, for instance, $\lambda^{\square}(p_+, p_-)$ is represented by

$$\begin{pmatrix} 0 & K^{\square}(\mid p_{+}, p_{-}) \\ -(K^{\square})^{t}(\mid p_{+}, p_{-}) & 0 \end{pmatrix}, \tag{8.26}$$

and similarly for the other partition distributions considered above.) We observe that for any set N, the Fredholm pfaffian

$$pf(J - \sqrt{t}K\sqrt{t})_N \tag{8.27}$$

encodes the distribution of $|\{\lambda_i - i\} \cap N\}|$, and thus as n varies,

$$pf(J - \sqrt{t}K\sqrt{t})_{\{n,n+1,\dots\}}$$
(8.28)

encodes the marginal distribution of λ_i for each i. With this in mind, we give the following Fredholm pfaffian identity:

Theorem 8.4. Let K be an antisymmetric matrix kernel that represents a probability distribution on the set of partitions. Then for any decomposition $\mathbb{Z} = N_+ \uplus N_-$ such that $N_{+-} := N_+ \cap \mathbb{Z}^-$ and $N_{-+} := N_- \cap \mathbb{N}$ are both finite,

$$pf(J - t^{1/4}(K - \chi_{N_{-}}J\chi_{N_{-}})t^{1/4})_{\mathbb{Z}} = (1 + \sqrt{t})^{|N_{-+}| - |N_{+-}|} pf(J - \sqrt{t}K\sqrt{t})_{N_{+}},$$
(8.29)

$$= (1 - \sqrt{t})^{|N_{+-}| - |N_{-+}|} \operatorname{pf}(J - \sqrt{t}(J - K)\sqrt{t})_{N_{-}}.$$
(8.30)

where $\chi_{N_{-}}$ is the projection onto N_{-} .

Proof. Let λ be the random partition associated to K, and set $T := \{\lambda_j - j\}$, $T_+ = T \cap N_+$, $T_- = N_- - T$. By the definition of the Fredholm pfaffian,

$$pf(J - t^{1/4}(K - \chi_{N_{-}}J\chi_{N_{-}})t^{1/4}) = \sum_{S \subset \mathbb{Z}} t^{|S|/2} pf((\chi_{N_{-}}J\chi_{N_{-}} - K)(S))$$
(8.31)

$$= \sum_{S_{\pm} \subset N_{\pm}} t^{(|S_{+}|+|S_{-}|)/2} \operatorname{pf} \begin{pmatrix} -K(S_{+}, S_{+}) & -K(S_{+}, S_{-}) \\ -K(S_{-}, S_{+}) & (J - K)(S_{-}, S_{-}) \end{pmatrix}$$
(8.32)

$$= \sum_{S_{\pm} \subset N_{\pm}} t^{(|S_{+}|+|S_{-}|)/2} (-1)^{|S_{+}|} \operatorname{pf} \begin{pmatrix} K(S_{+}, S_{+}) & \sqrt{-1}K(S_{+}, S_{-}) \\ \sqrt{-1}K(S_{-}, S_{+}) & (J - K)(S_{-}, S_{-}) \end{pmatrix}$$
(8.33)

$$= \sum_{S_{+} \subset N_{+}} t^{(|S_{+}| + |S_{-}|)/2} (-1)^{|S_{+}|} \Pr(S_{+} \subset T, S_{-} \cap T = \emptyset)$$
(8.34)

$$= \sum_{S_{\pm} \subset N_{\pm}} t^{(|S_{+}| + |S_{-}|)/2} (-1)^{|S_{+}|} \Pr(S_{\pm} \subset T_{\pm})$$
(8.35)

$$= \sum_{R_{\pm} \subset N_{\pm}} \sum_{S_{\pm} \subset R_{\pm}} t^{(|S_{+}| + |S_{-}|)/2} (-1)^{|S_{+}|} \Pr(T_{\pm} = R_{\pm})$$
(8.36)

$$= \sum_{R_{\pm} \subset N_{\pm}} (1 + \sqrt{t})^{|R_{-}|} (1 - \sqrt{t})^{|R_{+}|} \Pr(T_{\pm} = R_{\pm})$$
(8.37)

$$= \sum_{R_{+} \subset N_{+}} (1-t)^{|R_{+}|} (1+\sqrt{t})^{|R_{-}|-|R_{+}|} \Pr(T_{\pm} = R_{\pm}). \tag{8.38}$$

Now, we have the following lemma:

Lemma 8.5. Let $\mathbb{Z} = N_+ \cap N_-$ be a decomposition as above. Then for any partition λ with associated set T,

$$|N_{+} \cap T| - |N_{-} - T| = |N_{+-}| - |N_{-+}|. \tag{8.39}$$

Proof. Recall that for any partition,

$$|T \cap \mathbb{N}| = |\mathbb{Z}^- - T|. \tag{8.40}$$

Setting $N_{++} = N_+ \cap \mathbb{N}$, $N_{--} = N_- \cap \mathbb{Z}^-$, we have

$$|T \cap \mathbb{N}| = |N_{++} \cap T| + |N_{-+} \cap T| = |N_{++} \cap T| + |N_{-+}| - |N_{-+} - T| \tag{8.41}$$

and

$$|\mathbb{Z}^{-} - T| = |N_{+-} - T| + |N_{--} - T| = |N_{+-}| - |N_{+-} \cap T| + |N_{--} - T|. \tag{8.42}$$

Subtracting these two quantities, we conclude that

$$|N_{+} \cap T| + |N_{-+}| - |N_{-} - T| - |N_{+-}| = 0. \tag{8.43}$$

We may thus replace $(1+\sqrt{t})^{|R_-|-|R_+|}$ in the above sum with $(1+\sqrt{t})^{|N_-|-|N_+|}$. We thus have

$$pf(J + t^{1/4}(K - \chi_{N_{-}}J\chi_{N_{-}})t^{1/4}) = (1 + \sqrt{t})^{|N_{-+}| - |N_{+-}|} \sum_{R_{\pm} \subset N_{\pm}} (1 - t)^{|R_{+}|} Pr(T_{\pm} = R_{\pm})$$
(8.44)

$$= (1 + \sqrt{t})^{|N_{-+}| - |N_{+-}|} \sum_{R_{+} \subset N_{+}} (1 - t)^{|R_{+}|} \Pr(T_{+} = R_{+})$$
(8.45)

$$= (1 + \sqrt{t})^{|N_{-+}| - |N_{+-}|} \sum_{S_{+} \subset N_{+}} (-t)^{|S_{+}|} \Pr(S_{+} \subset T)$$
(8.46)

$$= \operatorname{pf}(J - \sqrt{t}K\sqrt{t})_{N_{+}}. \tag{8.47}$$

Similarly,

$$pf(J + t^{1/4}(K - \chi_{N_{-}}J\chi_{N_{-}})t^{1/4}) = (1 - \sqrt{t})^{|N_{+-}| - |N_{-+}|} \sum_{R_{-} \subset N_{-}} (1 - t)^{|R_{-}|} Pr(T_{-} = R_{-})$$
(8.48)

$$= (1 - \sqrt{t})^{|N_{+-}| - |N_{-+}|} \sum_{S_{-} \subset N_{-}} (-t)^{|S_{-}|} \Pr(S_{-} \cap T = \emptyset)$$
 (8.49)

$$= (1 - \sqrt{t})^{|N_{+-}| - |N_{-+}|} \operatorname{pf}(J - \sqrt{t}(J - K)\sqrt{t})_{N_{-}}.$$
(8.50)

Remark 1. The point of the theorem is that while

$$pf(J + t^{1/4}(K - \chi_{N_{-}}J\chi_{N_{-}})t^{1/4})_{\mathbb{Z}}$$
(8.51)

is rather more complicated as a pfaffian on \mathbb{Z} , its image under the Fourier transform (which as an orthogonal transformation preserves Fredholm pfaffians) is much more likely than

$$pf(J + t^{1/2}Kt^{1/2})_{N_{+}} (8.52)$$

to have a simple kernel on the unit circle. Indeed, for the first pfaffian to have a simple kernel, all that is necessary is for K and χ_{N_-} to have simple kernels; for the second pfaffian, their composition must also be simple.

Remark 2. Note that in particular,

$$pf(J - \sqrt{t}(J - K)\sqrt{t})_{N_{-}} = (1 - t)^{|N_{-+}| - |N_{+-}|} pf(J - \sqrt{t}K\sqrt{t})_{N_{+}}.$$
(8.53)

Corollary 8.6. Let K be a scalar kernel such that

$$\begin{pmatrix} 0 & K \\ -K^t & 0 \end{pmatrix} \tag{8.54}$$

represents a probability distribution on the set of partitions. Then for any decomposition $\mathbb{Z} = N_+ \oplus N_-$ such that $N_{+-} := N_+ \cap \mathbb{Z}^-$ and $N_{-+} := N_- \cap \mathbb{N}$ are both finite,

$$\det(I - t^{1/2}(K - \chi_{N_{-}}))_{\mathbb{Z}} = (1 + \sqrt{t})^{|N_{-+}| - |N_{+-}|} \det(I - tK)_{N_{+}}, \tag{8.55}$$

$$= (1 - \sqrt{t})^{|N_{+-}| - |N_{-+}|} \det(I - t(I - K))_{N_{-}}. \tag{8.56}$$

For instance, taking $K = K^{\square}(||p_+, p_-|)$ and conjugating by the Fourier transform, we find

$$\det(1 - \lambda K)_{[n,\infty)} = (1 + \sqrt{\lambda})^{-n} \det(I - \lambda^{1/2} K')_C, \tag{8.57}$$

where

$$K'(z,w) = \frac{z^{-n}w^n - \phi(z)\phi(w)^{-1}}{2\pi i(z-w)},$$
(8.58)

$$\phi(z) = \frac{E(z; p_+)}{E(1/z; p_-)},\tag{8.59}$$

and with C an appropriately chosen contour containing 0. This generalizes the results of [1] (which essentially showed that when $p_+ = p_- = t$:/, the identity holds to second order at $\lambda = 1$). For a direct, analytic proof of this identity, see [2].

We close by remarking that [6] used the identity of [11] to express a large class of Toeplitz determinants as discrete Fredholm determinants, or equivalently, to so express a large class of integrals over the unitary group. Similarly, Corollaries 4.3 and 5.2 can be used to express appropriate integrals over the orthogonal and symplectic groups as discrete Fredholm pfaffians:

$$\int_{U \in O(l)} \det(E(U; p)) = Z^{\square}(p; 0)^{-1} \operatorname{pf}(J - K^{\square'}(\mid p; 0))_{[l, \infty)}$$
(8.60)

$$\int_{U \in Sp(2l)} \det(E(U;p)) = Z^{\Sigma}(p;0)^{-1} \operatorname{pf}(J - K^{\Sigma'}(\mid p;0))_{[2l,\infty)}$$
(8.61)

(actually statements about formal integrals); here

$$Z^{\boxtimes}(p;0) := \operatorname{pf}(J - K^{\boxtimes'}(\mid p;0))_{[0,\infty)}$$
(8.62)

$$Z^{\mathbb{N}}(p;0) := \operatorname{pf}(J - K^{\mathbb{N}'}(\mid p;0))_{[0,\infty)}. \tag{8.63}$$

We can also use Theorem 8.4 to rewrite these as continuous Fredholm pfaffians; details are left to the reader.

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